

# First Order Logic

ITI0210, lecture 8 (2021)

# Review

Entailment:  $KB \models \alpha$

using existing **knowledge** to deduce e.g. answers to questions

Truth value algorithms:

- truth table model checking
- DPLL

Syntax based inference:

- Forward chaining on Horn clauses
- Resolution

# Motivation

In first order logic (FOL):

$$Pit(X) \wedge NextTo(Y, X) \rightarrow Breeze(Y)$$

In propositional logic:

$$Pit_{1,1} \rightarrow Breeze_{2,1}$$

$$Pit_{1,1} \rightarrow Breeze_{1,2}$$

$$Pit_{2,1} \rightarrow Breeze_{1,1}$$

$$Pit_{2,1} \rightarrow Breeze_{2,2}$$

$$Pit_{2,1} \rightarrow Breeze_{3,1}$$

etc...

# Motivation

In first order logic (FOL):

$$\textit{Brother}(Y, X) \rightarrow \textit{Male}(Y)$$

In propositional logic, not really possible

(unless we pick a subset of people and generate rules for them)

This type of **generalization** is key in A.I.



# FOL Syntax and Semantics

An informal introduction

# Some Examples

$Bird(tweety)$  - Tweety is a bird

$Bird(X) \rightarrow Flies(X)$  – If X is a bird, it flies

$\forall X Bird(X) \rightarrow Flies(X)$  – same thing more formally

$\exists X AtStanford(X) \wedge Student(X) \wedge Smart(X)$  – there is a student at Stanford who is smart

# Some Examples

*Bird*(*tweety*) - Tweety is a bird

*Bird*(*X*)  $\rightarrow$  *Flies*(*X*) – If X is a bird, it flies

$\forall X$  *Bird*(*X*)  $\rightarrow$  *Flies*(*X*) – same thing more formally

$\exists X$  *AtStanford*(*X*)  $\wedge$  *Student*(*X*)  $\wedge$  *Smart*(*X*) – there is a student at Stanford who is smart

Predicate

Constant

Variable

Quantifier

# More Examples

Multiple variables

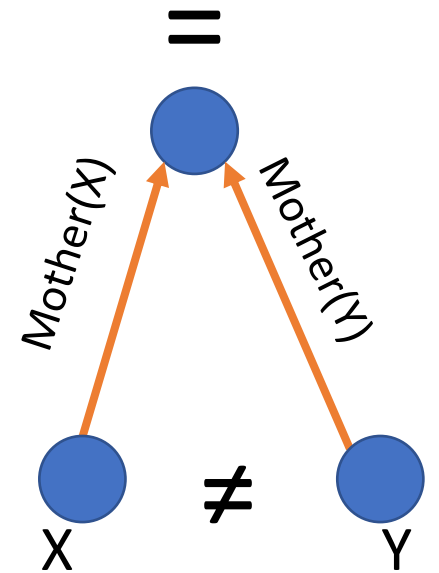
$$\textit{Smokes}(X) \wedge \textit{Influences}(X, Y) \rightarrow \textit{Smokes}(Y)$$

Functions

$$\textit{Human}(X) \rightarrow \textit{Human}(\textit{Mother}(X))$$

Equality

$$\textit{Girl}(X) \wedge \textit{Girl}(Y) \wedge \textit{Mother}(X) = \textit{Mother}(Y) \wedge \neg(X = Y) \rightarrow \textit{Sister}(X, Y)$$





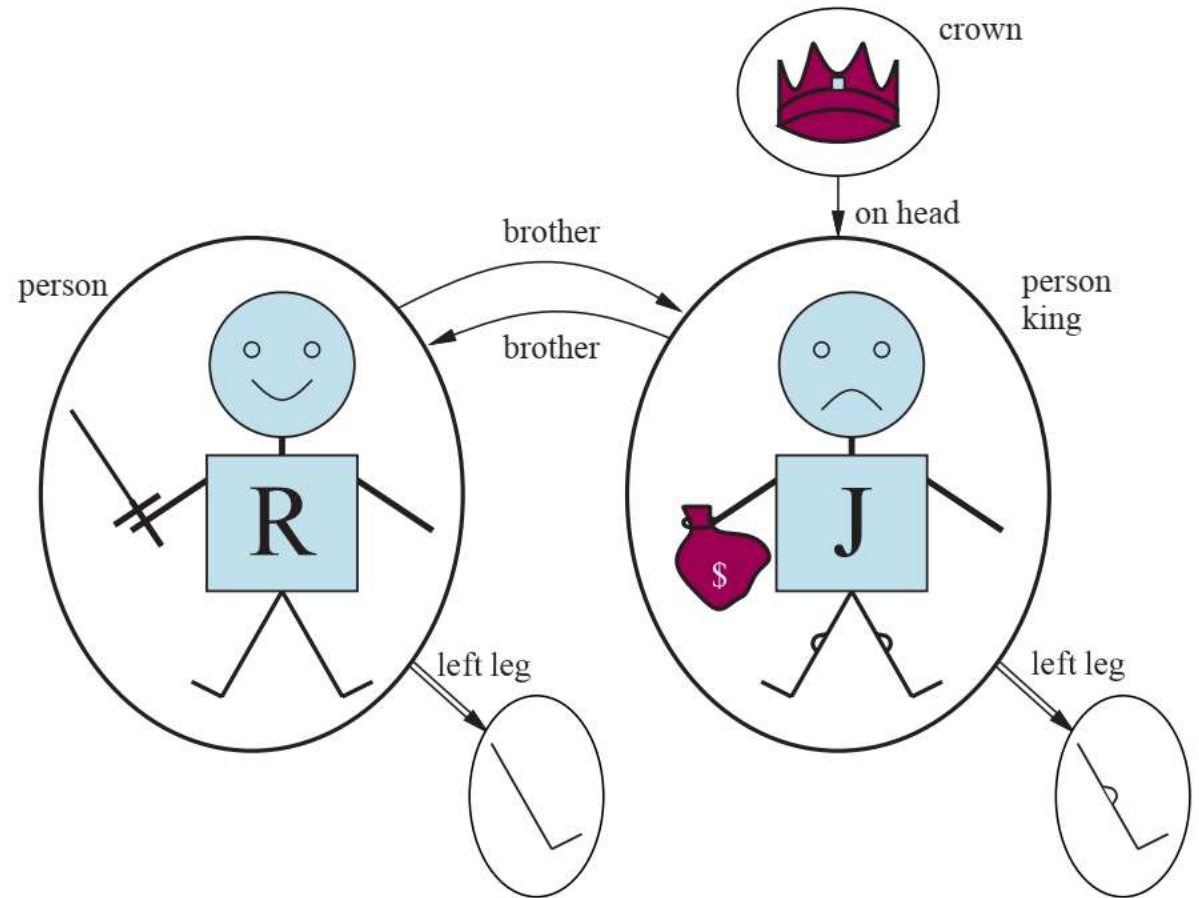
# Semantics

Semantics, or meaning of sentences/formulas

Everything refers to an object, relation, function



Each sentence has a truth value (like in propositional logic)



# Semantics

facts:

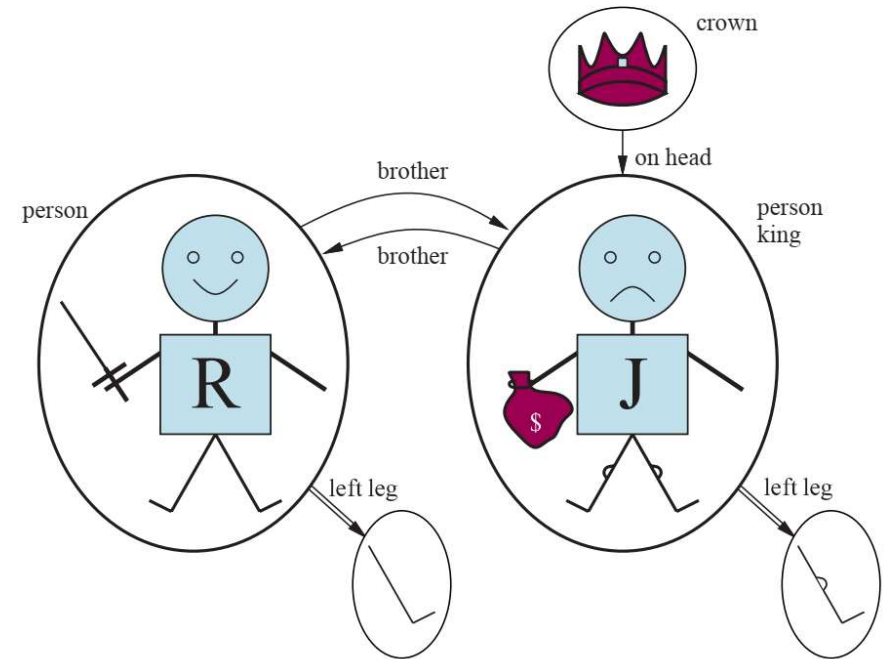
*Person(richard)*  
*Person(john)*  
*Brother(richard, john)*  
*Brother(john, richard)*  
*King(john)*

reasoning about royal headgear:

$King(X) \wedge Onhead(X, Y) \rightarrow Crown(Y)$

each brother has their own left leg:

$Brother(X, Y) \rightarrow \neg(LeftLeg(X) = LeftLeg(Y))$




Semantics: each predicate is about some relation, each constant/variable/function about an object

# Semantics

Each formula has a **truth value**, because FOL formulas consist of:

Terms:  $Function(term_1, \dots, term_n)$  or *constant* or *variable*

Atomic sentence:  $Predicate(term_1, \dots, term_n)$

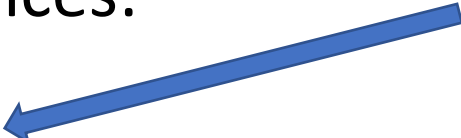


Predicates are  
always True  
or False

Construct recursively from atomic sentences:

if  $S_1$  and  $S_2$  are sentences, then

$\neg S_1, S_1 \vee S_2, S_1 \wedge S_2, S_1 \rightarrow S_2, S_1 \Leftrightarrow S_2$  are sentences



Can compute truth  
values as each  $S$  is  
True/False

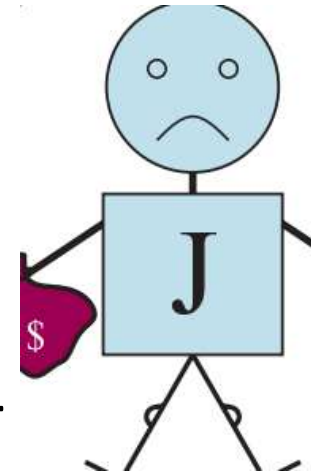
# Semantics

Fun fact about semantics:

The reasoner has no idea about the intended semantics

The constant “richard” could mean this guy too

This is why model checking is not possible in FOL  
(Can't test every possible value of symbols)

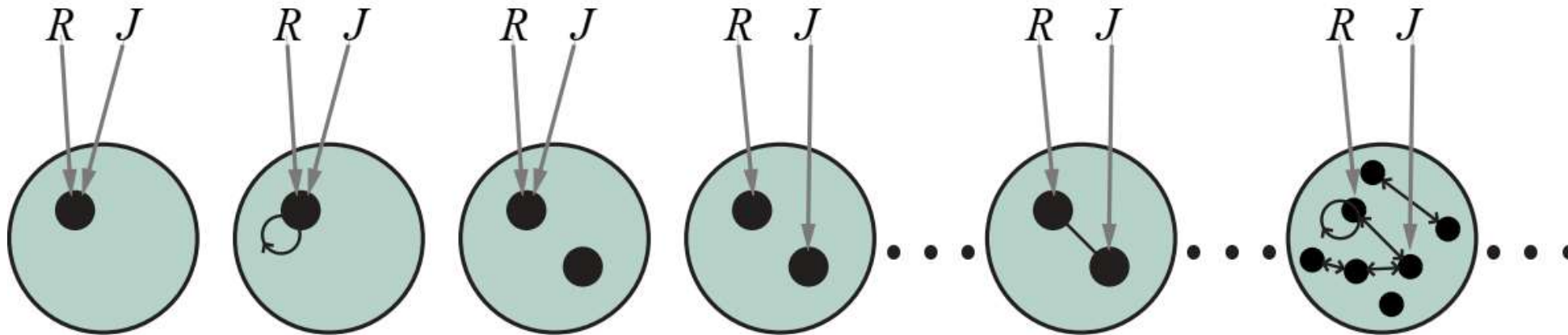


# FOL Inference

Deducing things from FOL knowledge

# Deciding Entailment

Generally, not possible to decide  $KB \models \alpha$   
(too many/infinite number of models to check)



Pictured: models for two constant symbols and one relation

# Usable Inference Methods

High end theorem provers:

- full first order logic
- resolution

PROLOG:

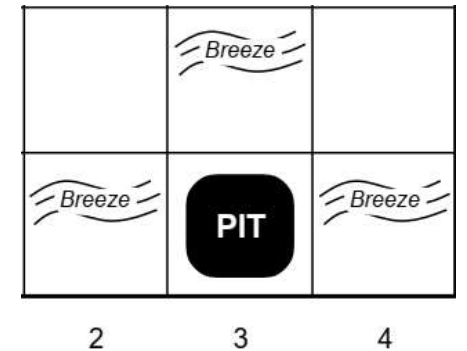
- simplified language
- depth first search

# Resolution in FOL

Handle variables and constants by **unification**:

$$\frac{\neg Pit(X) \vee Dangerous(X) \quad Pit(square_{3,1})}{Dangerous(square_{3,1})}$$

(substitution  $\{X/square_{3,1}\}$  applied to input clauses)



Accessible explanation in Estonian (not part of the course):

Tamme, T., Tammet, T., Prank, R. Loogika: mõtlemisest tõestamiseni, TÜ Kirjastus, 1997

<https://dspace.ut.ee/handle/10062/24397>



# Knowledge Representation

The main use for FOL in AI

# Facts and Rules

“Grounded” facts, about specific entities

*Smokes(bob)*  
*Influencer(bob)*  
*Roommate(bob, carl)*

“Lifted” rules, allow generalized reasoning

$Smokes(X) \wedge Influencer(X) \wedge Roommate(X, Y) \rightarrow Smokes(Y)$

# Quantifiers

You probably noted chronic lack of  $\forall, \exists$  quantifiers on the slides

$$\textit{Bird}(X) \rightarrow \textit{Flies}(X)$$

$\forall X$  is implied here (the common sense way to read the rule)

The existential quantifier cannot be omitted

$$\exists X \textit{Loves}(X, \textit{alice})$$

But, an anonymous constant will encode the same knowledge

$$\textit{Loves}(\textit{anon1}, \textit{alice})$$

# Implementations

FOL:

[https://en.wikipedia.org/wiki/Otter \(theorem prover\)](https://en.wikipedia.org/wiki/Otter_(theorem_prover))

<https://github.com/tammet/gkc>

<https://vprover.github.io/>

PROLOG:

<http://www.gprolog.org/>

<https://www.swi-prolog.org/>

Each has **different syntax, capabilities** etc.

Lots of books about PROLOG!

# Open World

Test this input using <http://logictools.org/>

```
-innocent(X) => guilty(X).  
innocent(bob).  
% negated query  
guilty(bob).
```

What happens? What did you expect?

# Open World

What about this:

```
earthquake(X) => alarm(X).  
burglary(X) => alarm(X).  
-burglary(today).  
-earthquake(today).  
alarm(today).  
% or try this  
% -alarm(today).
```

Not able to prove “Some statement”  
does **not** mean “Some statement is false”

# Open World

## Gandalf revisited

**Wizards** wear a wizard hat  
**Gandalf** wears a wizard hat

WizardHat(gandalf).  
Wizard(X) => WizardHat(X).  
Wizard(X) => \$ans(X).



Turns out Gandalf wasn't a wizard after all?